



University of Information Technology – Vietnam National University HCMC
Faculty of Computer Science



CS431.N21 – Deep Learning and Applications

Analysis of FCN for Semantic Segmentation

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Introduction



Image Segmentation is a broad part of Computer Vision, in image segmentation we classify every pixel of the image into one of the classes.





Semantic segmentation



Instance segmentation

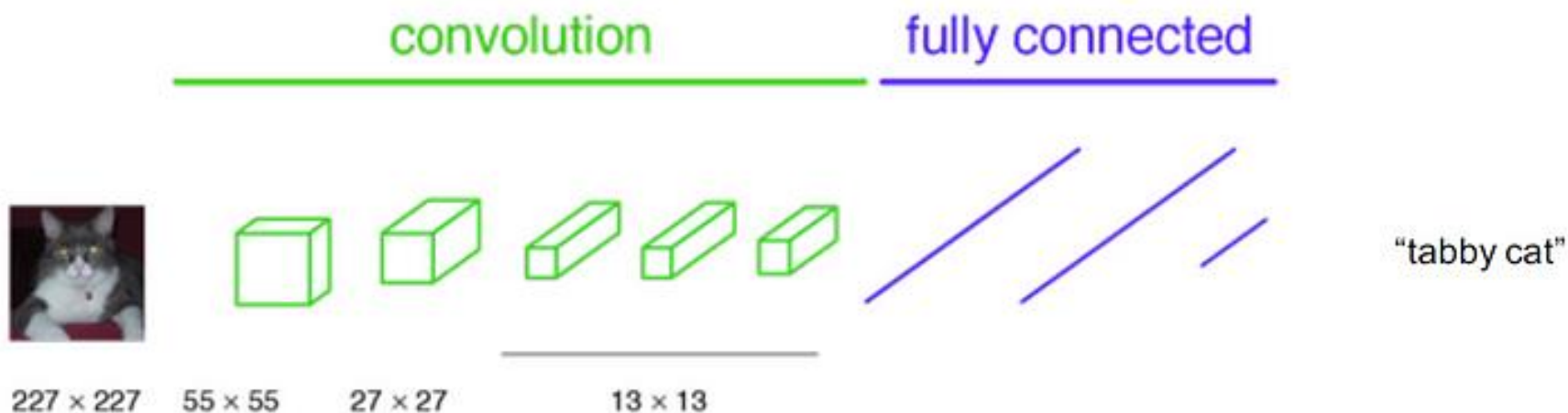




Fully Convolutional Networks (FCN)



Most of us may familiar with CNNs in Image Classification



Architecture of a CNN model



What if...?

convolution



227×227



55×55



27×27



13×13



1×1



CNN - From Image Classification to Semantic Segmentation



convolution



$H \times W$



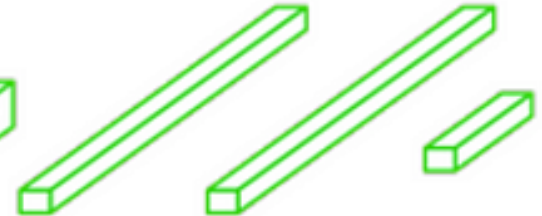
$H/4 \times W/4$



$H/8 \times W/8$



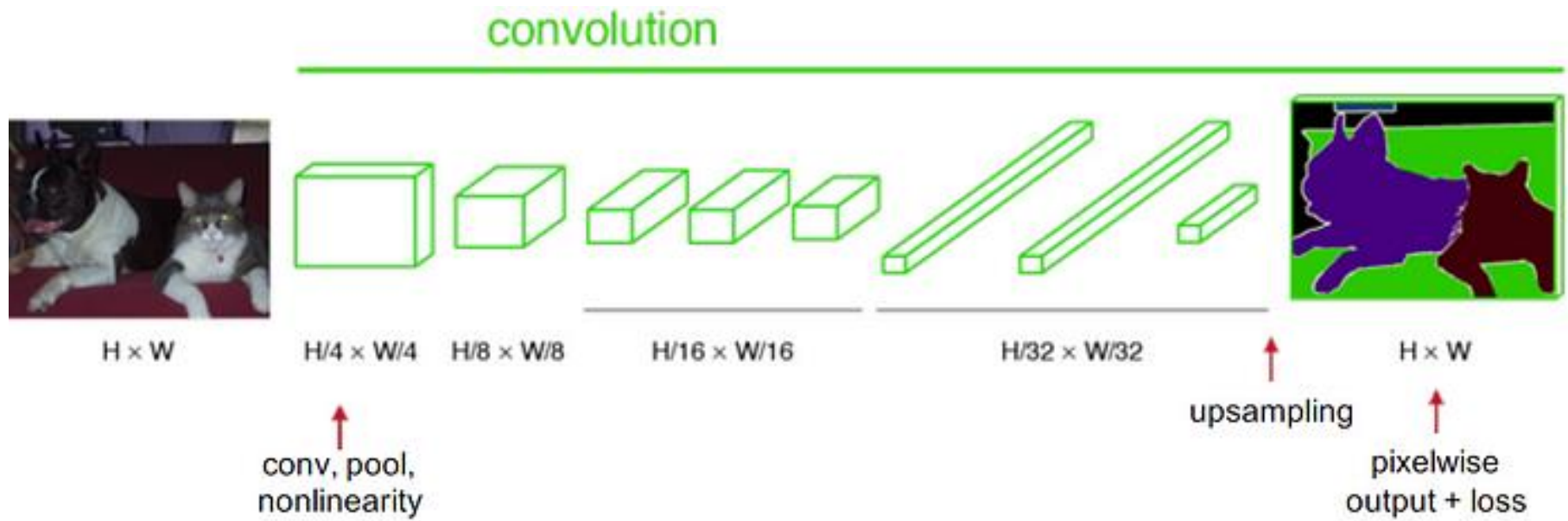
$H/16 \times W/16$



$H/32 \times W/32$



CNN - From Image Classification to Semantic Segmentation





Convolution Matrices are used to extract spatial features

-1	-1	-1	-1	-1	-1	-1	-1	-1
-1	1	-1	-1	-1	-1	-1	1	-1
-1	-1	1	-1	-1	-1	1	-1	-1
-1	-1	-1	1	-1	1	-1	-1	-1
-1	-1	-1	-1	1	-1	-1	-1	-1
-1	-1	-1	1	-1	1	-1	-1	-1
-1	-1	1	-1	-1	-1	1	-1	-1
-1	1	-1	-1	-1	-1	-1	1	-1
-1	-1	-1	-1	-1	-1	-1	-1	-1

 \otimes

1	-1	-1
-1	1	-1
-1	-1	1

 $=$

0.77	-0.11	0.11	0.33	0.55	-0.11	0.33
-0.11	1.00	-0.11	0.33	-0.11	0.11	-0.11
0.11	-0.11	1.00	-0.33	0.11	-0.11	0.55
0.33	0.33	-0.33	0.55	-0.33	0.33	0.33
0.55	-0.11	0.11	-0.33	1.00	-0.11	0.11
-0.11	0.11	-0.11	0.33	-0.11	1.00	-0.11
0.33	-0.11	0.55	0.33	0.11	-0.11	0.77



Convolution Layer



0	0	0	0	0	0	...
0	156	155	156	158	158	...
0	153	154	157	159	159	...
0	149	151	155	158	159	...
0	146	146	149	153	158	...
0	145	143	143	148	158	...
...

Input Channel #1 (Red)

0	0	0	0	0	0	...
0	167	166	167	169	169	...
0	164	165	168	170	170	...
0	160	162	166	169	170	...
0	156	156	159	163	168	...
0	155	153	153	158	168	...
...

Input Channel #2 (Green)

0	0	0	0	0	0	...
0	163	162	163	165	165	...
0	160	161	164	166	166	...
0	156	158	162	165	166	...
0	155	155	158	162	167	...
0	154	152	152	157	167	...
...

Input Channel #3 (Blue)

-1	-1	1
0	1	-1
0	1	1

Kernel Channel #1



308

1	0	0
1	-1	-1
1	0	-1

Kernel Channel #2



-498

0	1	1
0	1	0
1	-1	1

Kernel Channel #3



164

+

+



Bias = 1

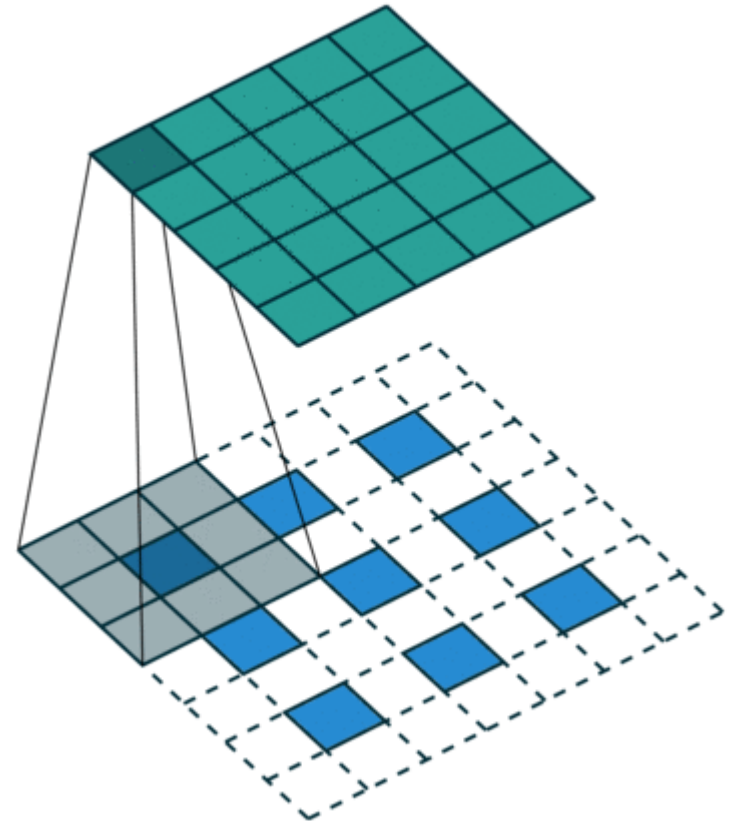
+ 1 = -25

Output

-25				...
				...
				...
				...
...

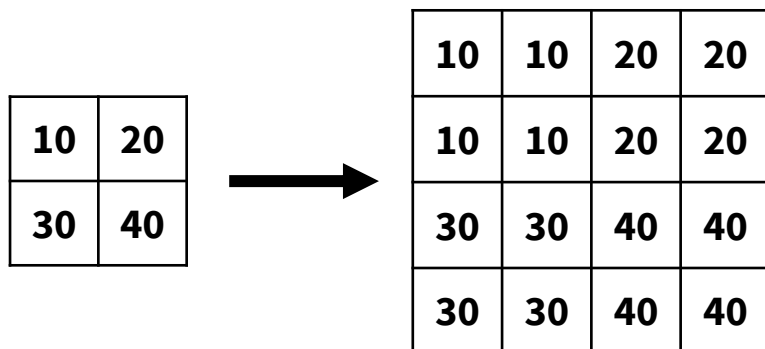


DeConvolution is a technique to upsample the image by using Convolution Matrices.

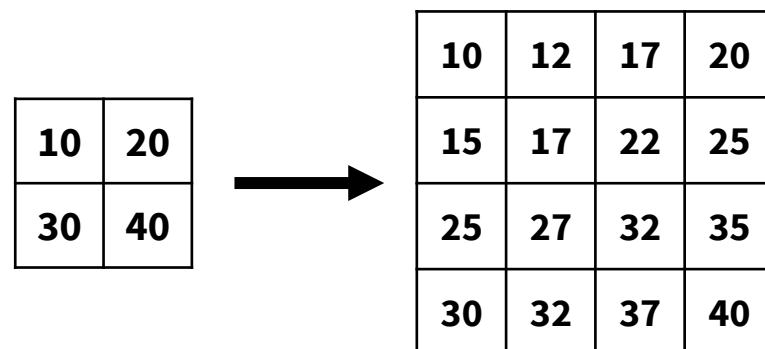




Why DeConvolution ?

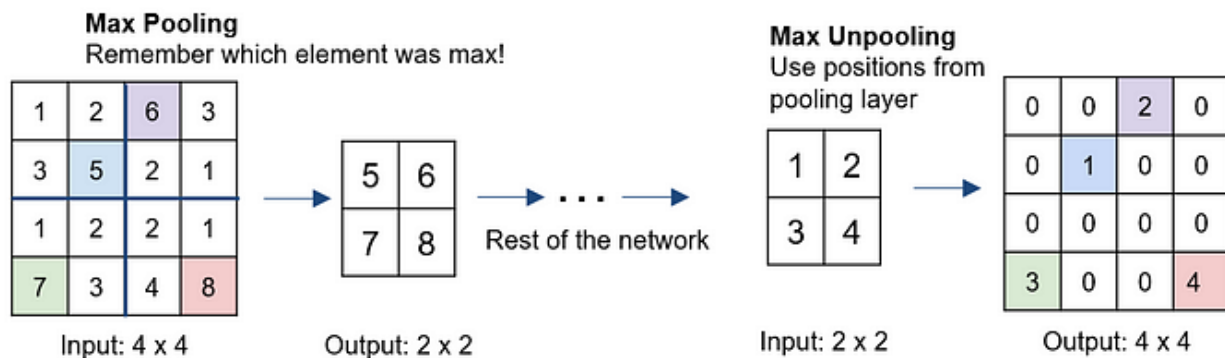


Nearest Neighbor

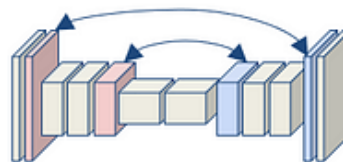


Bi-Linear Interpolation

Unpooling



Corresponding pairs of
downsampling and
upsampling layers





Problems with these upscaling techniques ?

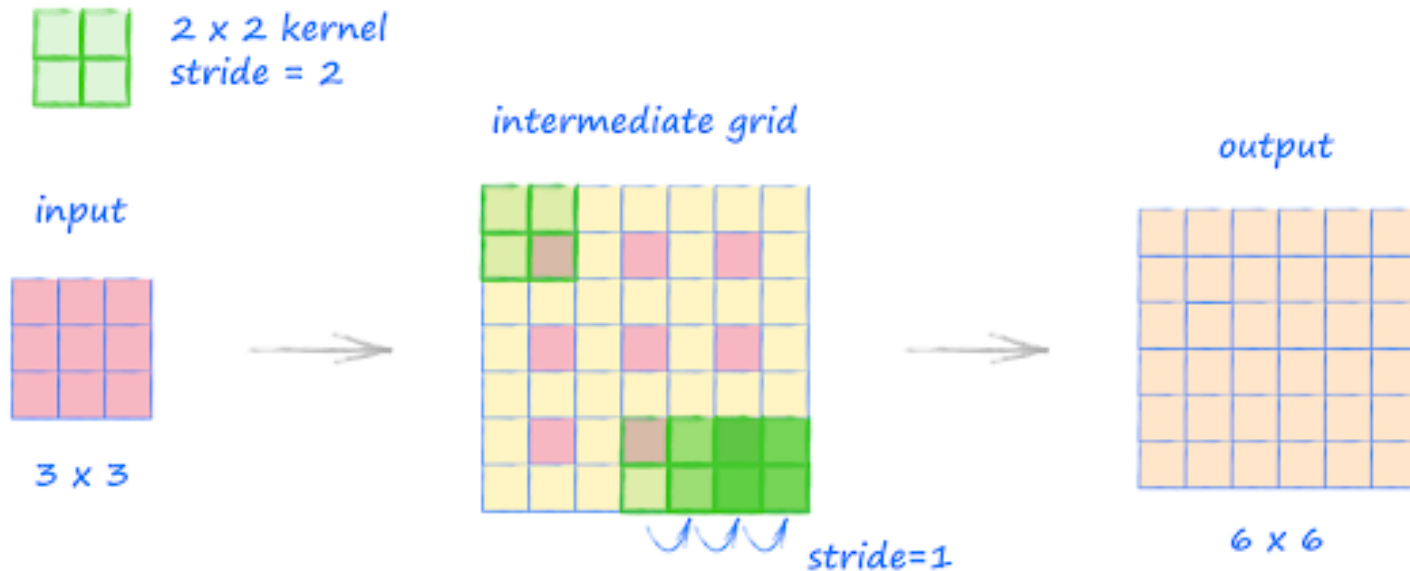
They do not learn from data



DeConvolution - Procedure



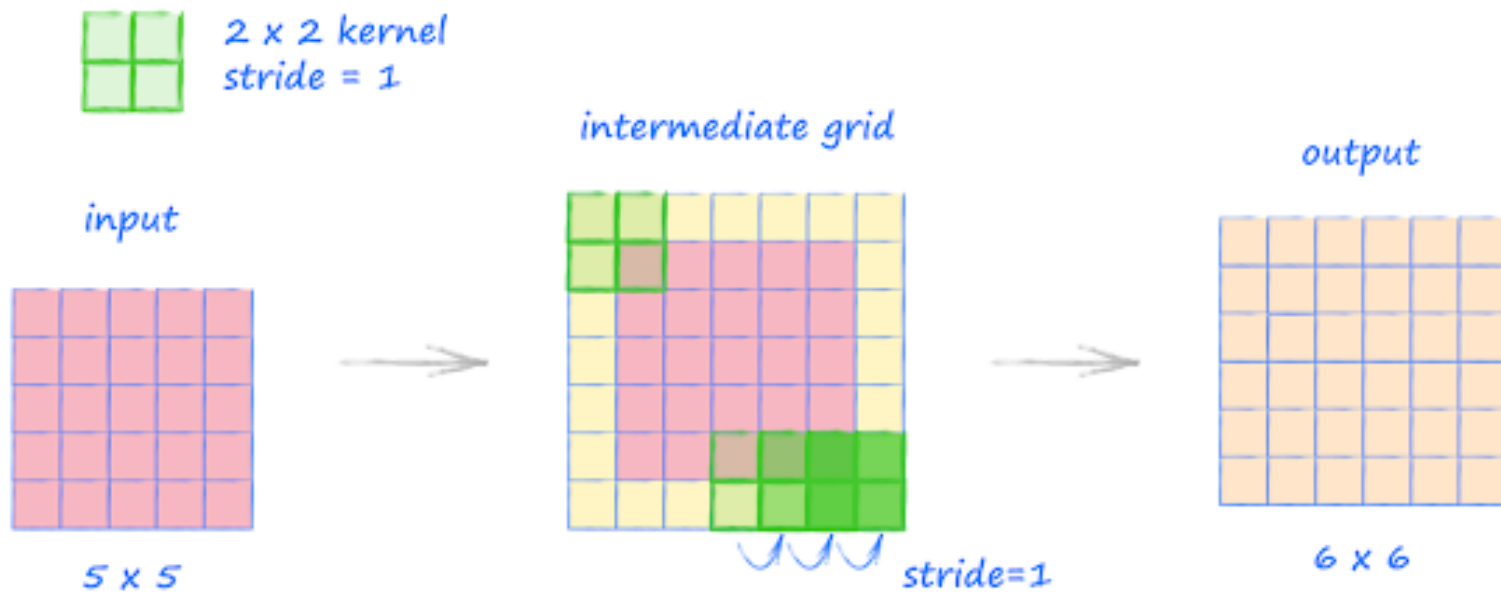
- **Step 1:** Apply the transformation to input (Add 0-cells between rows and cols)
- **Step 2:** Do convolution to input with given padding and stride informations



Stride 2, No Padding



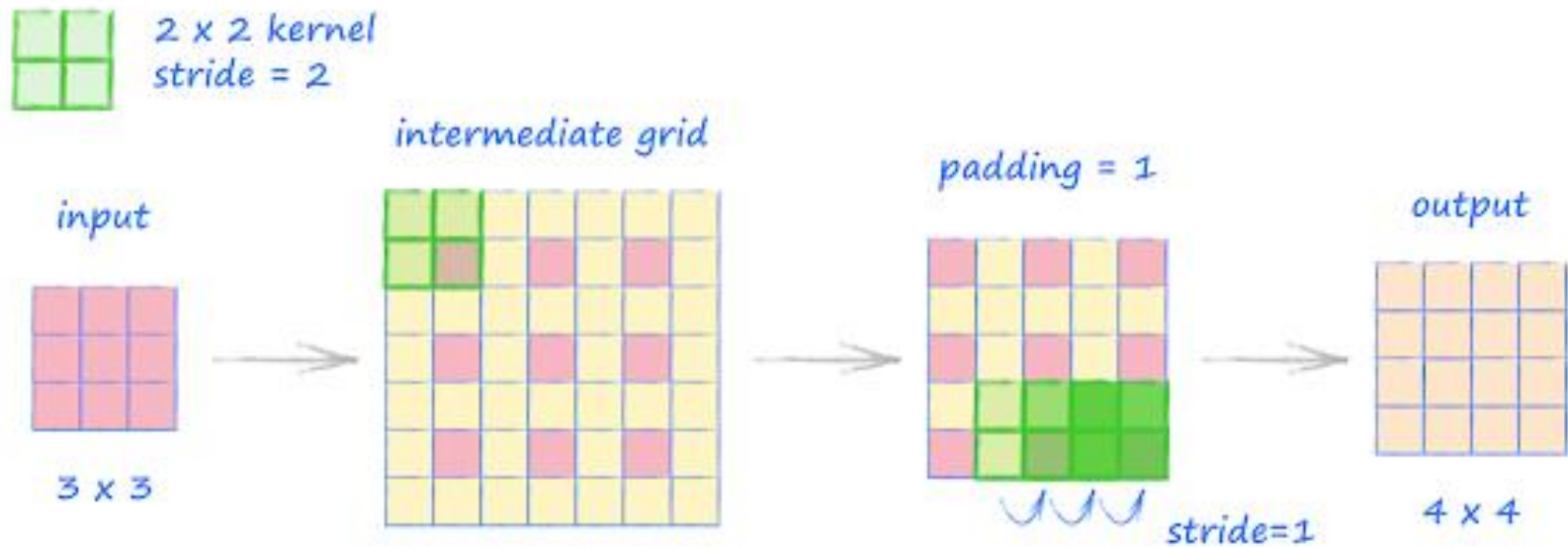
DeConvolution – Other examples



Stride 1, No Padding



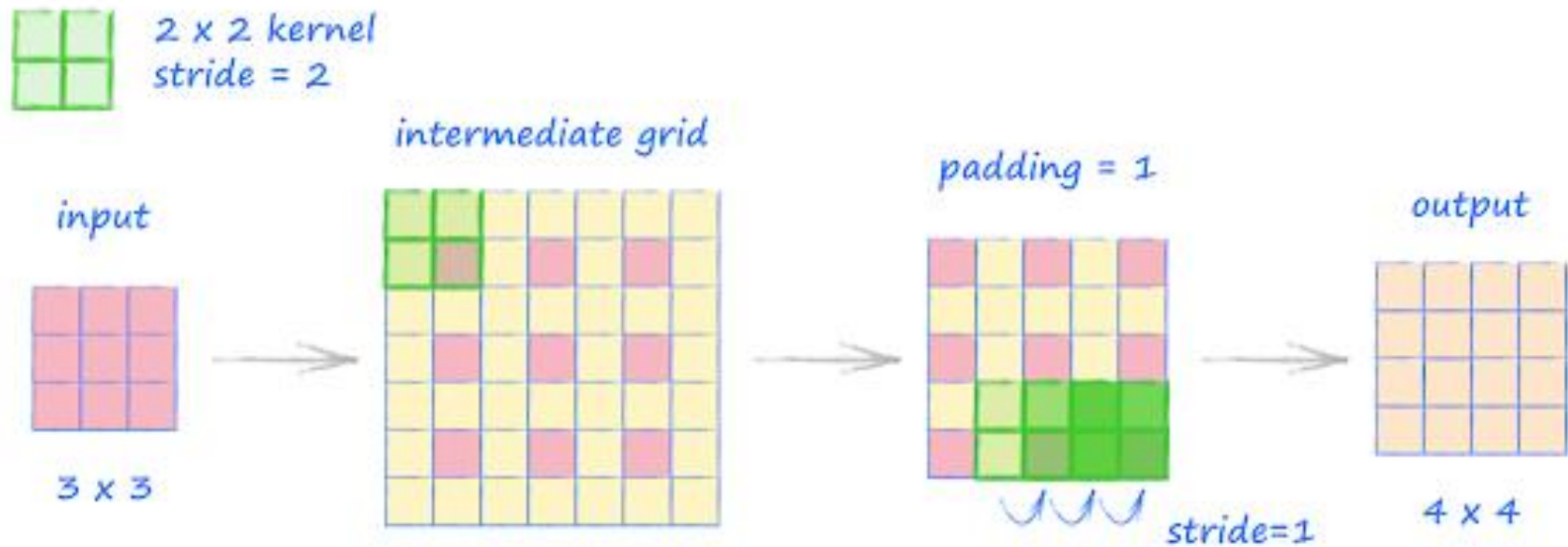
DeConvolution – Other examples



Stride 2, Padding



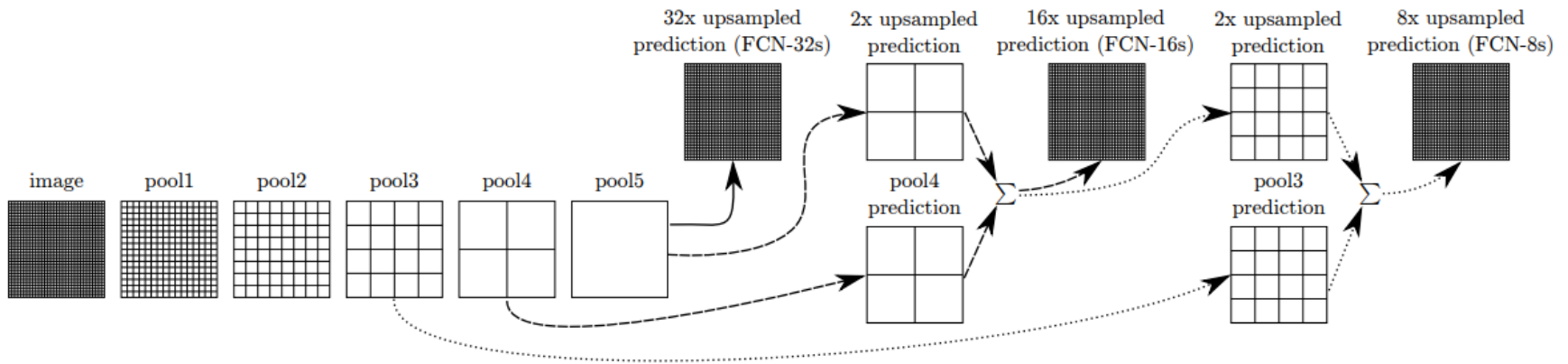
DeConvolution – Other examples



Stride 2, Padding



Architecture of FCN (32s/16s/8s)

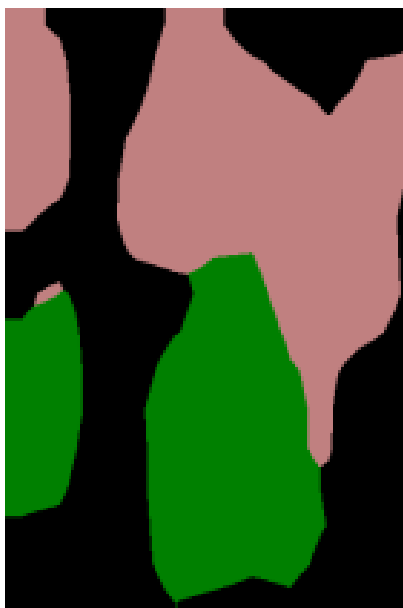




Comparison between FCNs



FCN-32s



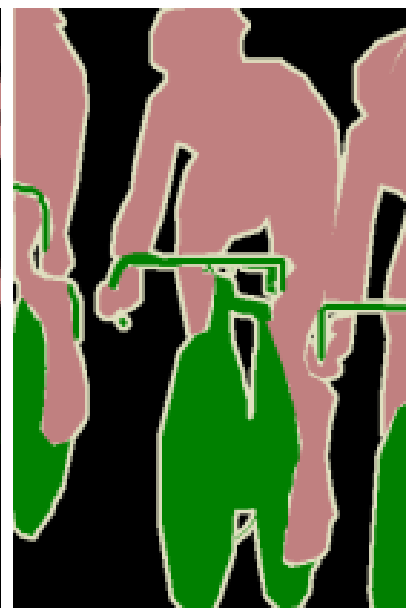
FCN-16s



FCN-8s



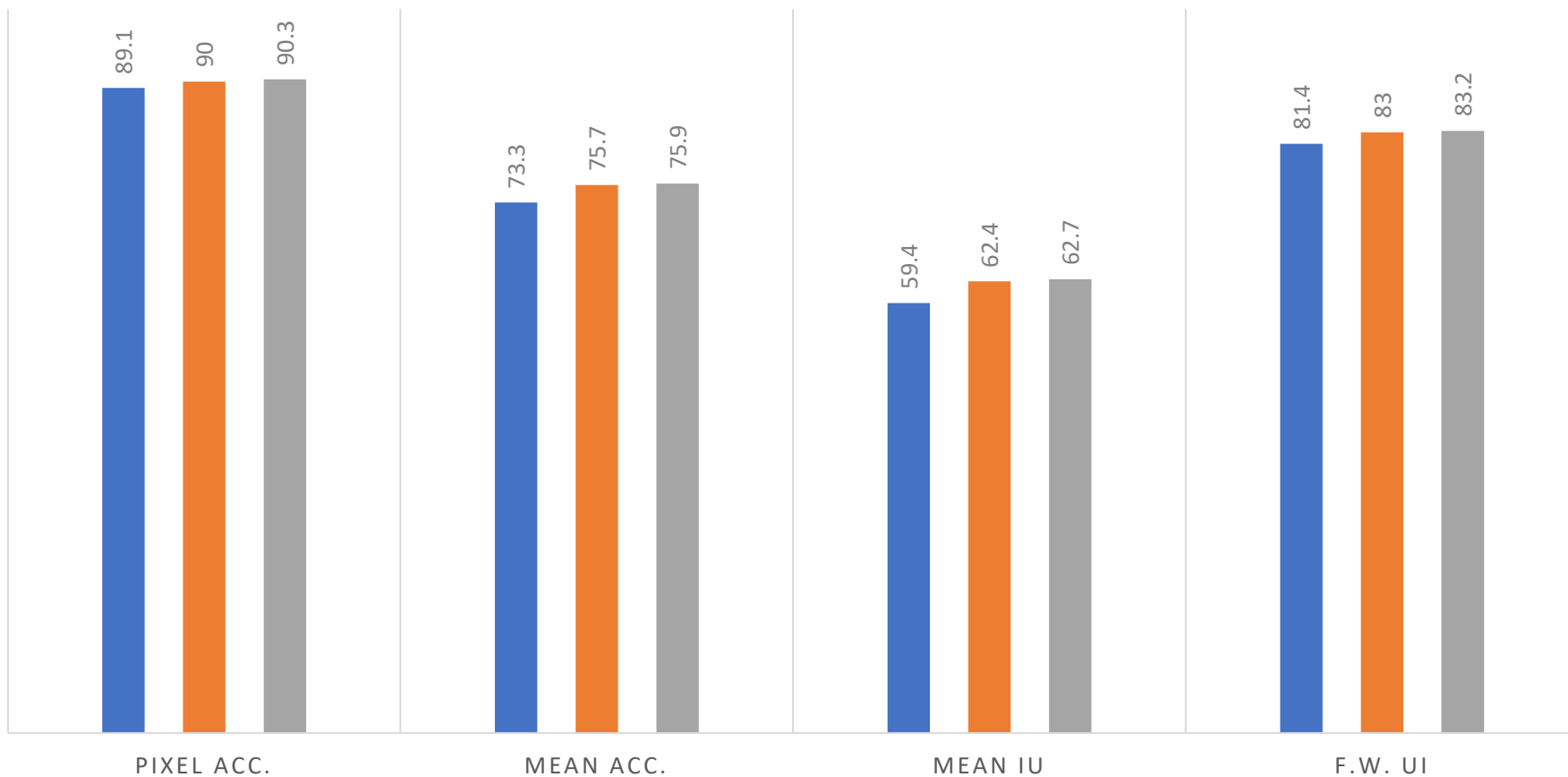
Ground truth





COMPARISON ON PASCAL VOC 2011

FCN-32s FCN-16s FCN-8s



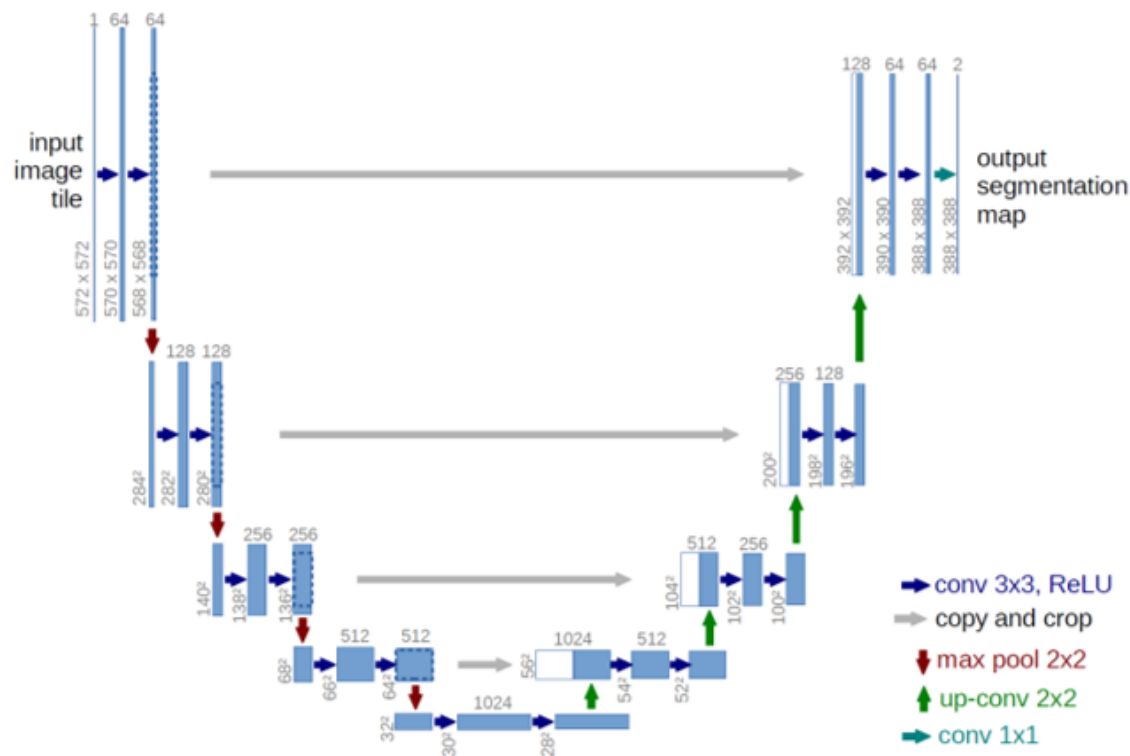


U-Net



U-Net is a convolutional neural network originally developed for segmenting biomedical images. When visualized its architecture looks like the letter U and hence the name U-Net.

Architecture

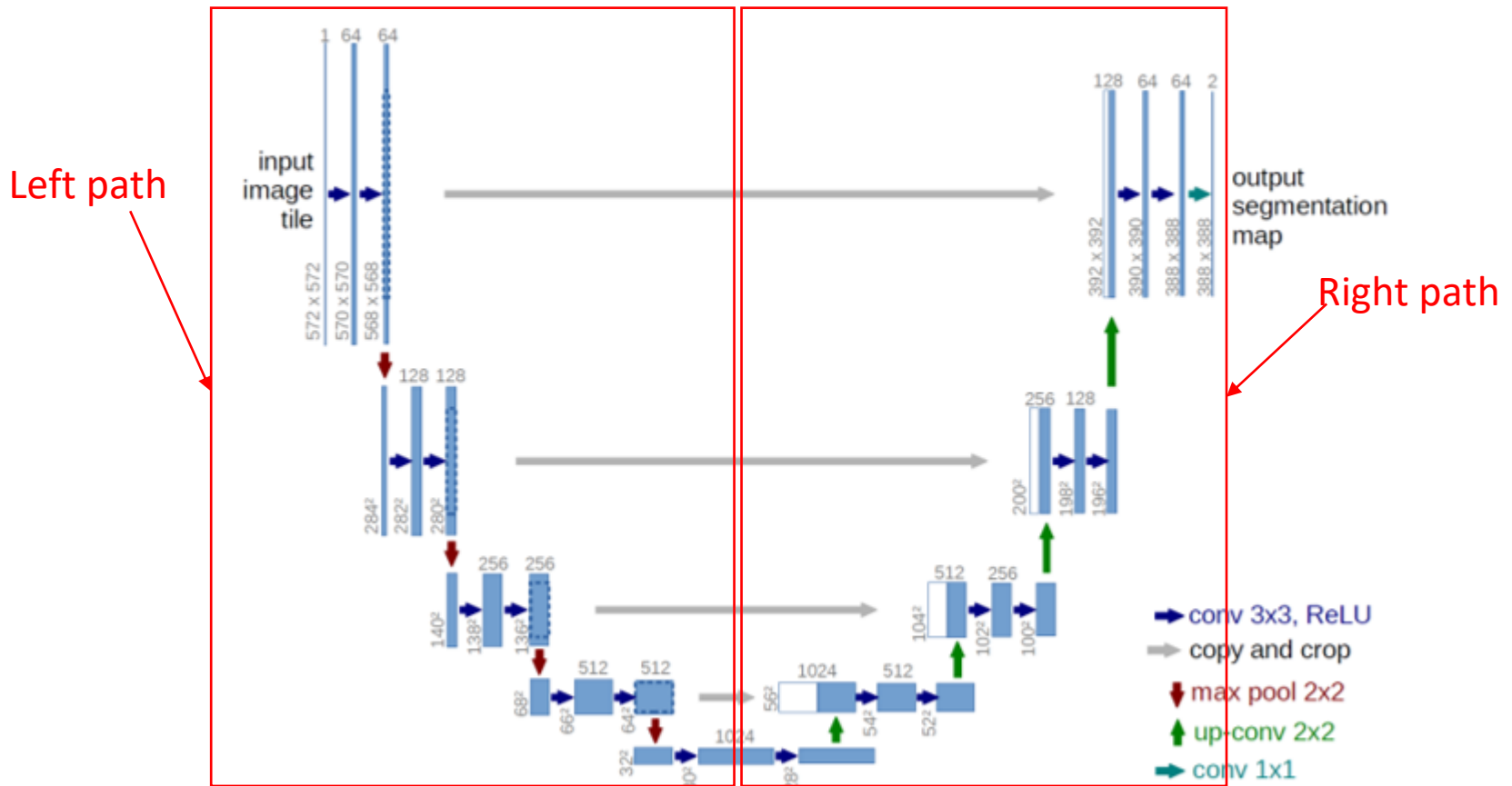




Unet's architecture is made up of two parts:

- The left path: contracting path
- The right path: expansive path

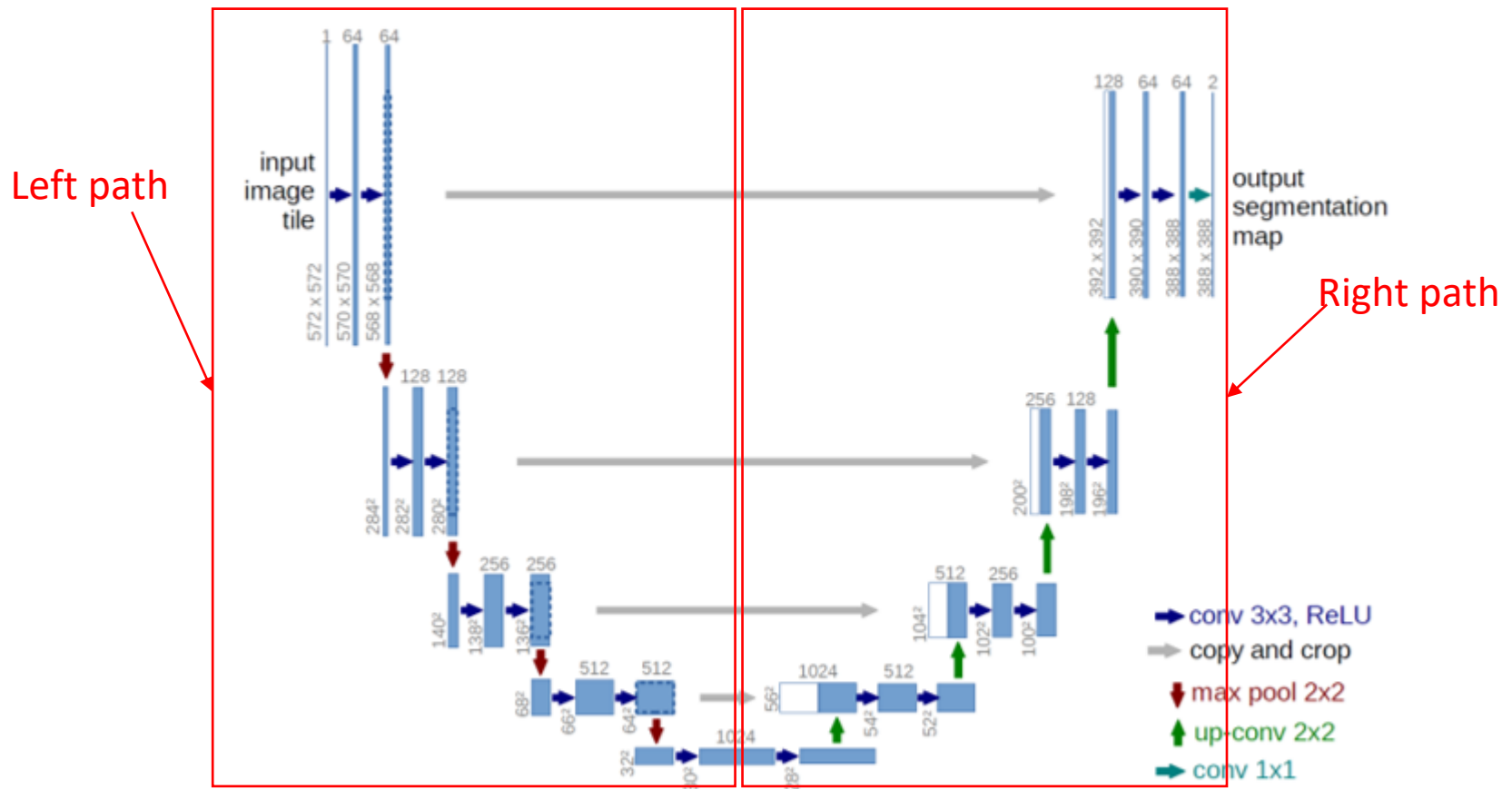
Architecture





- Contracting path: capture context
- Expansive path: aid in precise localization

Architecture



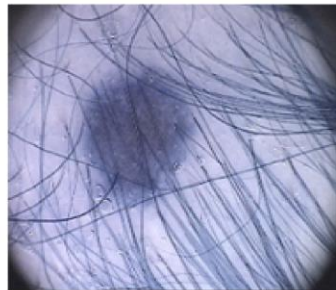
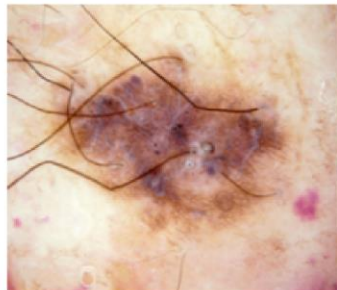
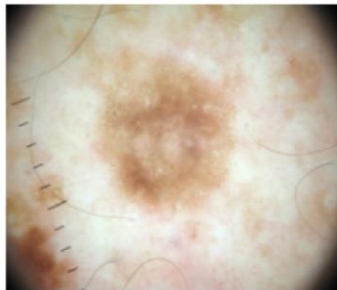


- With Unet in segmentation, we need to reconstruct the image from the feature vector created by CNN.
- We convert feature map into a vector and reconstruct an image from this vector.
- Unet has no any fully connected layers. The features are connected by the right path, so it don't need fully connected network.





- Medical imaging
- Object detection
- Traffic control systems
- Machine vision
- Content-based image retrieval





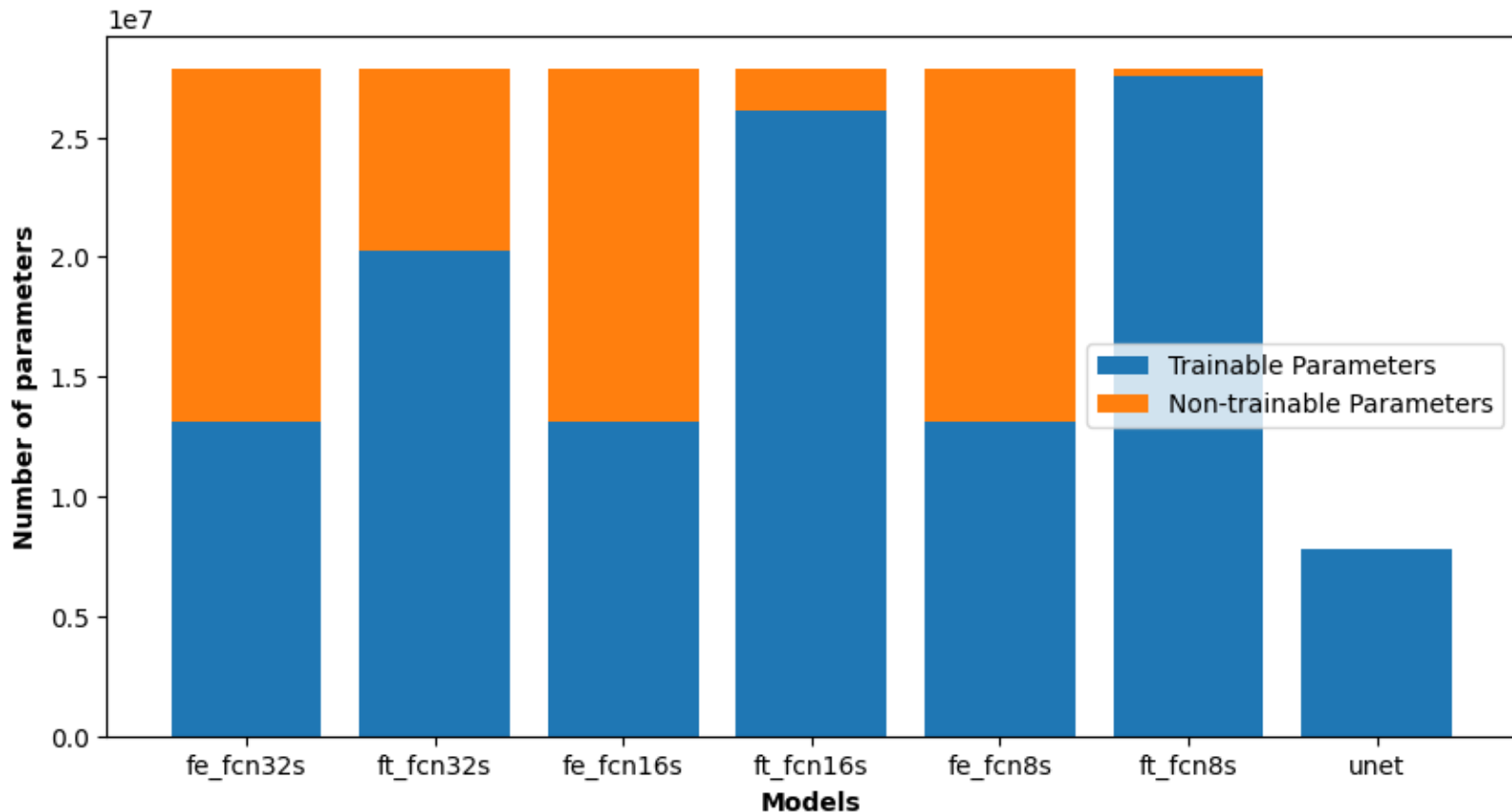
Experiments



Total parameters

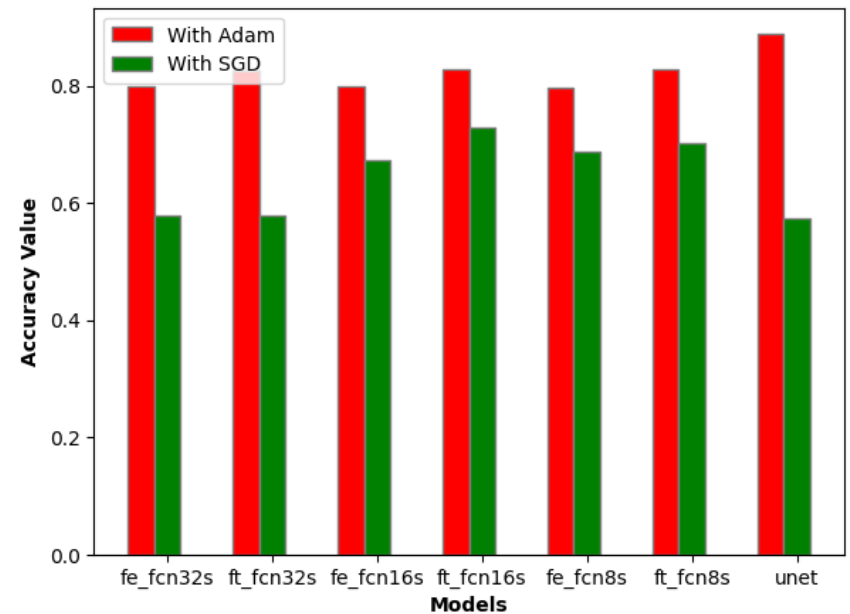
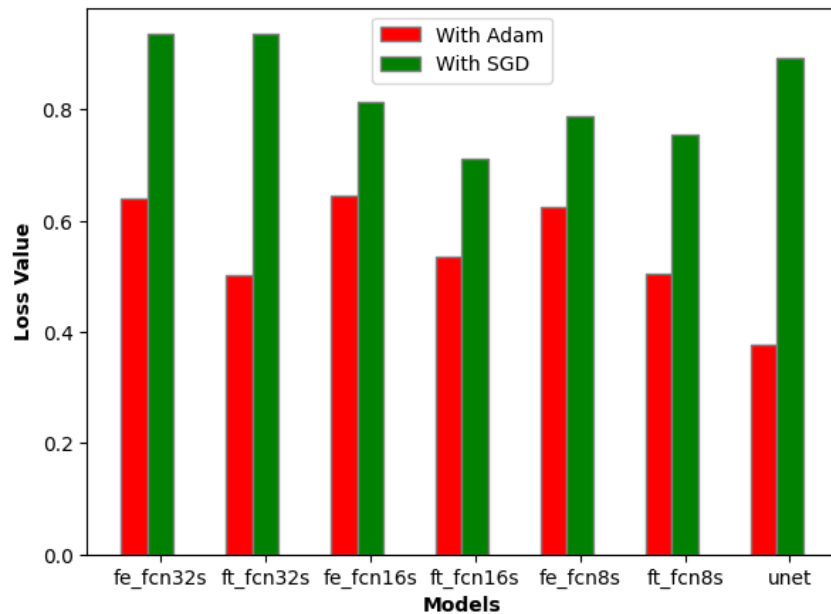


- The comparison about total parameters among the types of models





- The comparison of loss and accuracy

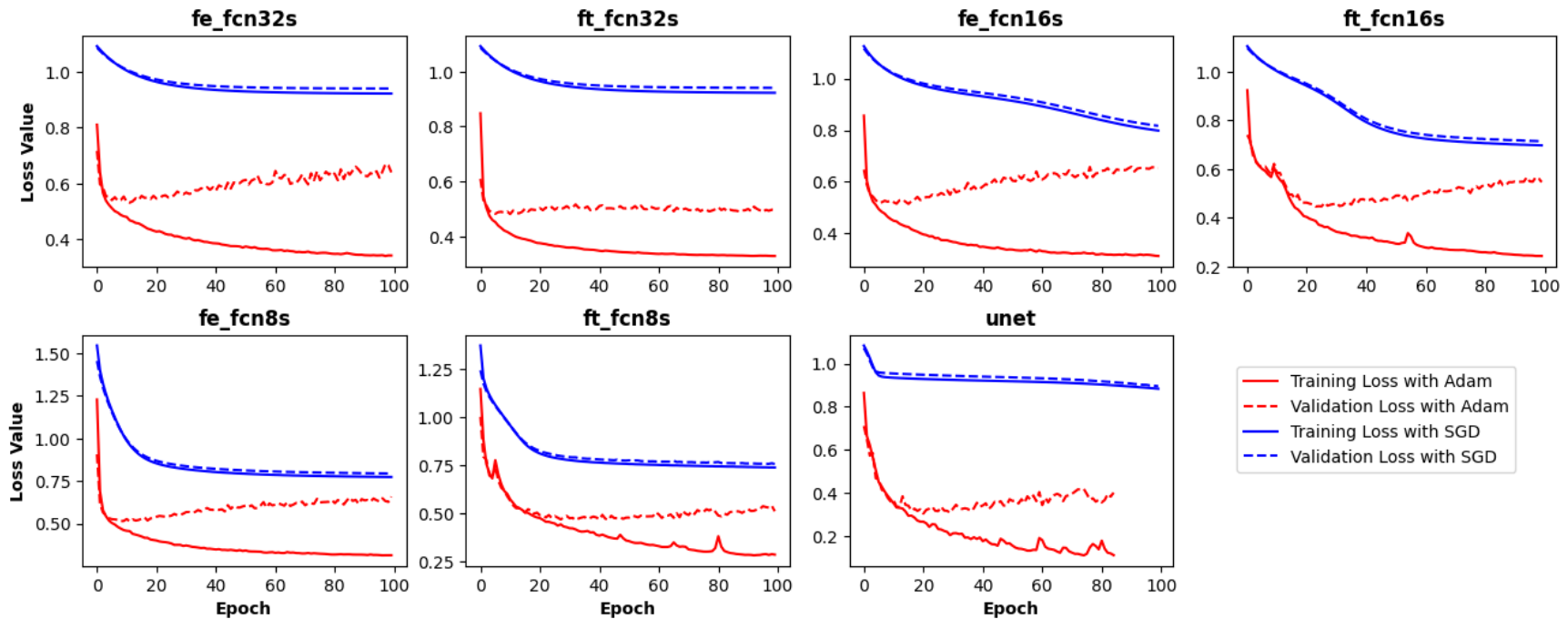




Loss convergence speed

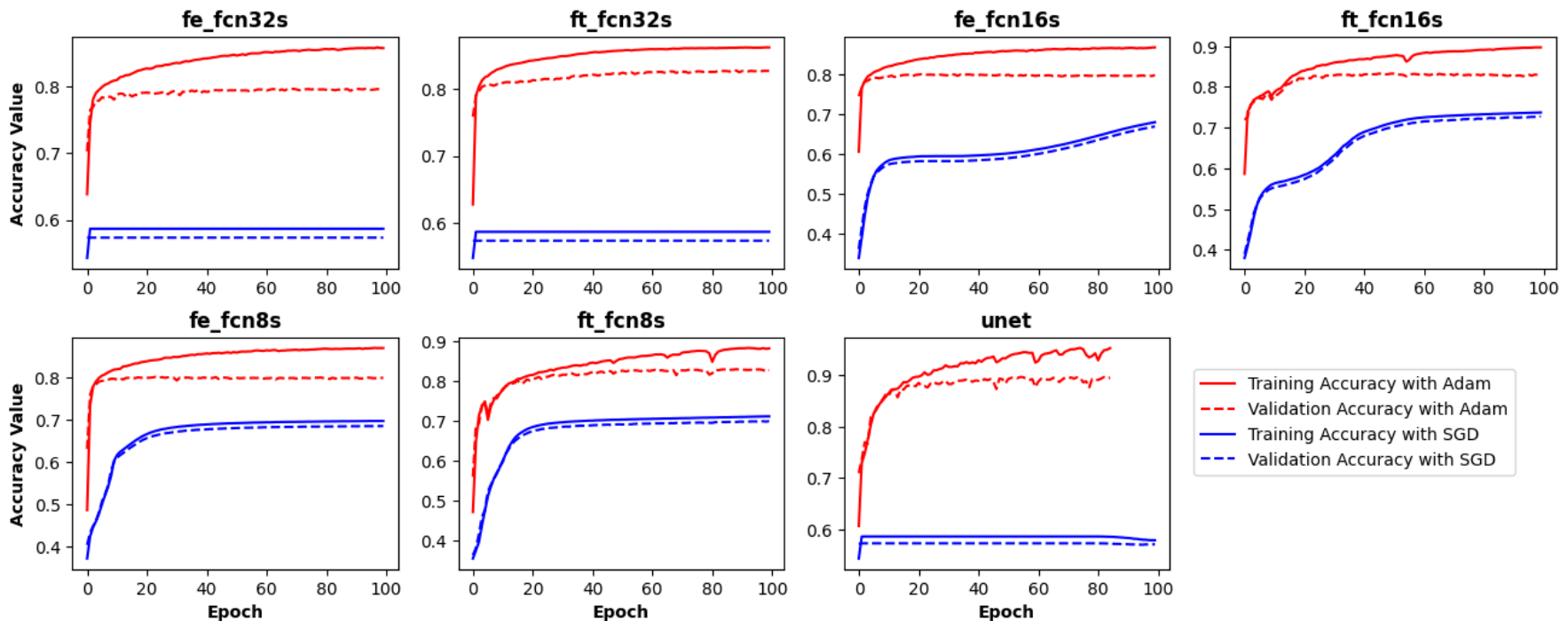


- The comparison of loss convergence speed



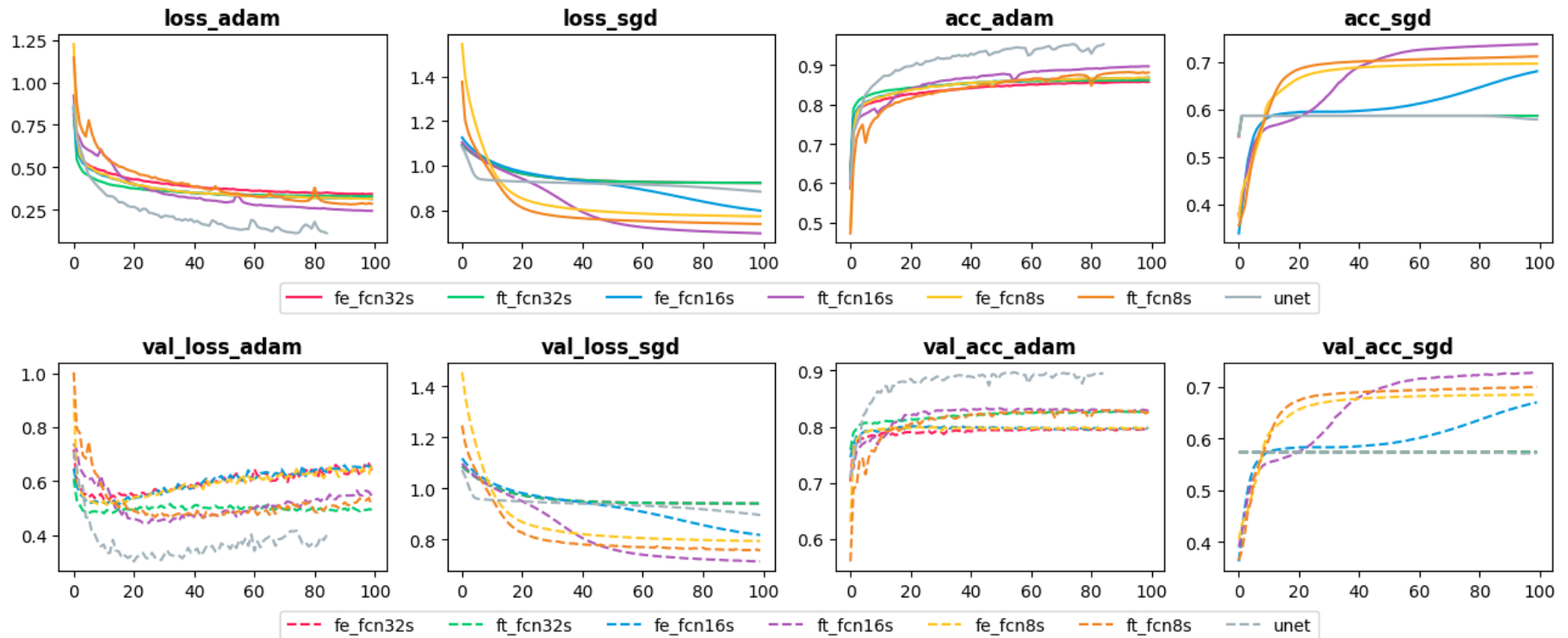


- The comparison of accuracy convergence speed





- The comparison of all convergences



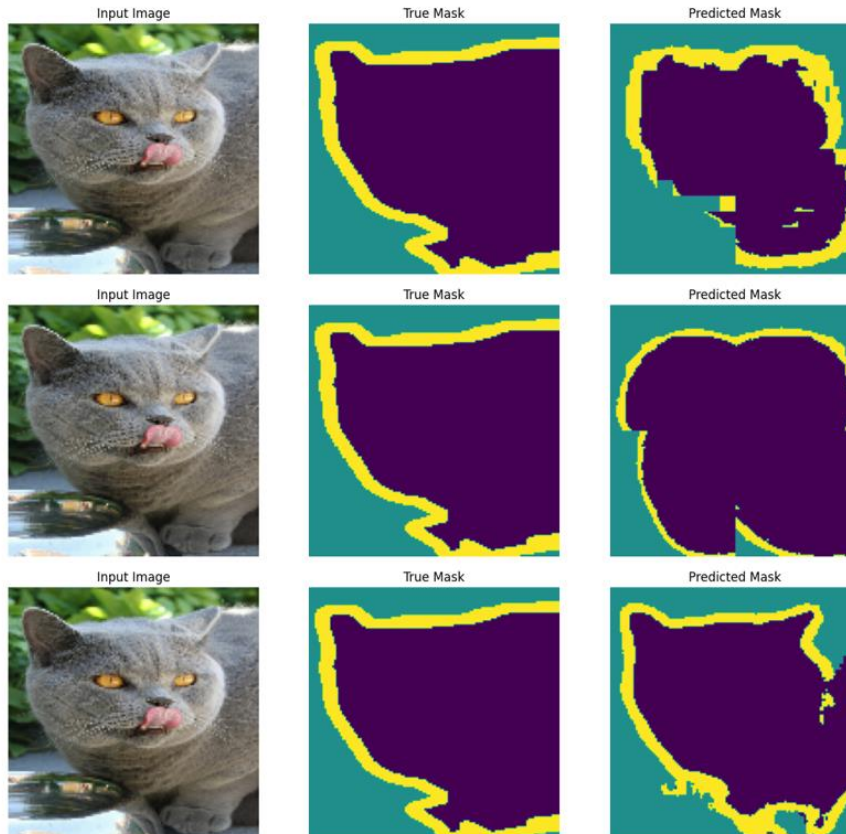


Comparison the predictions



- The comparison of predictions

7



ft_fcn16s

fe_fcn16s

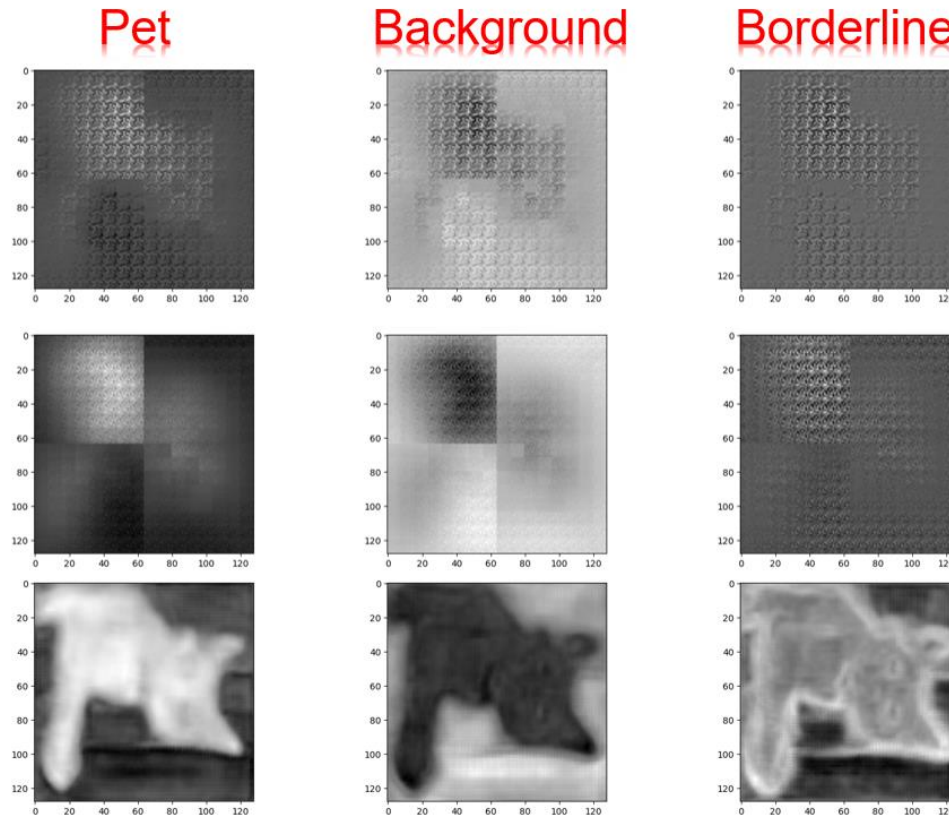
Unet



Comparison the feature maps



- The comparison of feature maps



ft_fcn16s

fe_fcn16s

Unet



Unet & FCN Demo:

<https://colab.research.google.com/drive/1iisL6plYsJCXE21KPNvDZD5uQucmX1tt?usp=sharing>

https://colab.research.google.com/drive/1e_55b_8aiam9h_P-NPuNkQXZL_LZGsZq

<https://colab.research.google.com/drive/1DGWbaS2lQcp1tAXizUQHVPsrMmkCvL2U?usp=sharing>

[Squeeze U-Net: A Memory and Energy Efficient Image Segmentation Network \(thecvf.com\)](#)

[\[1505.04597\] U-Net: Convolutional Networks for Biomedical Image Segmentation \(arxiv.org\)](#)